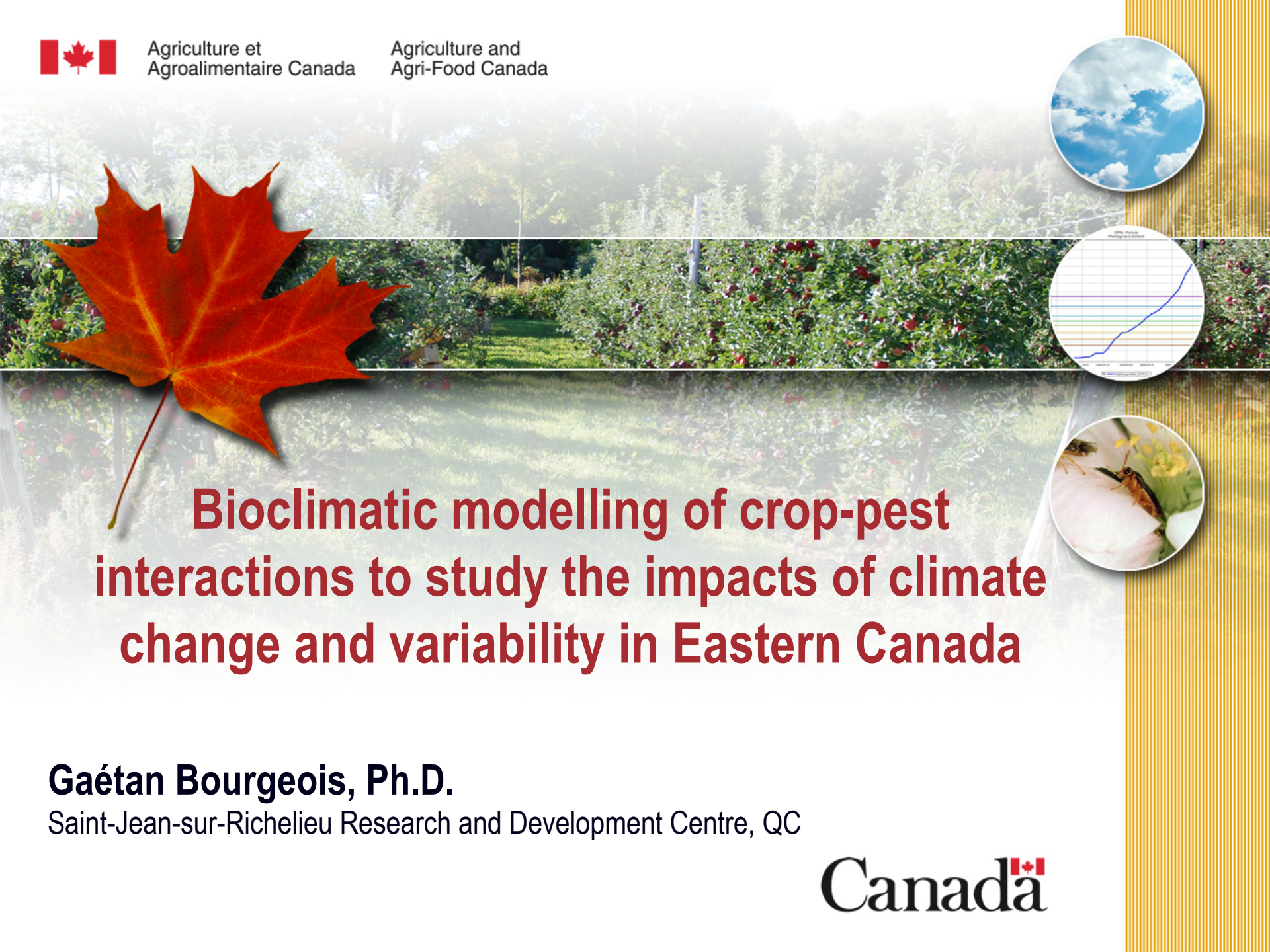

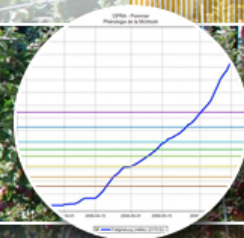




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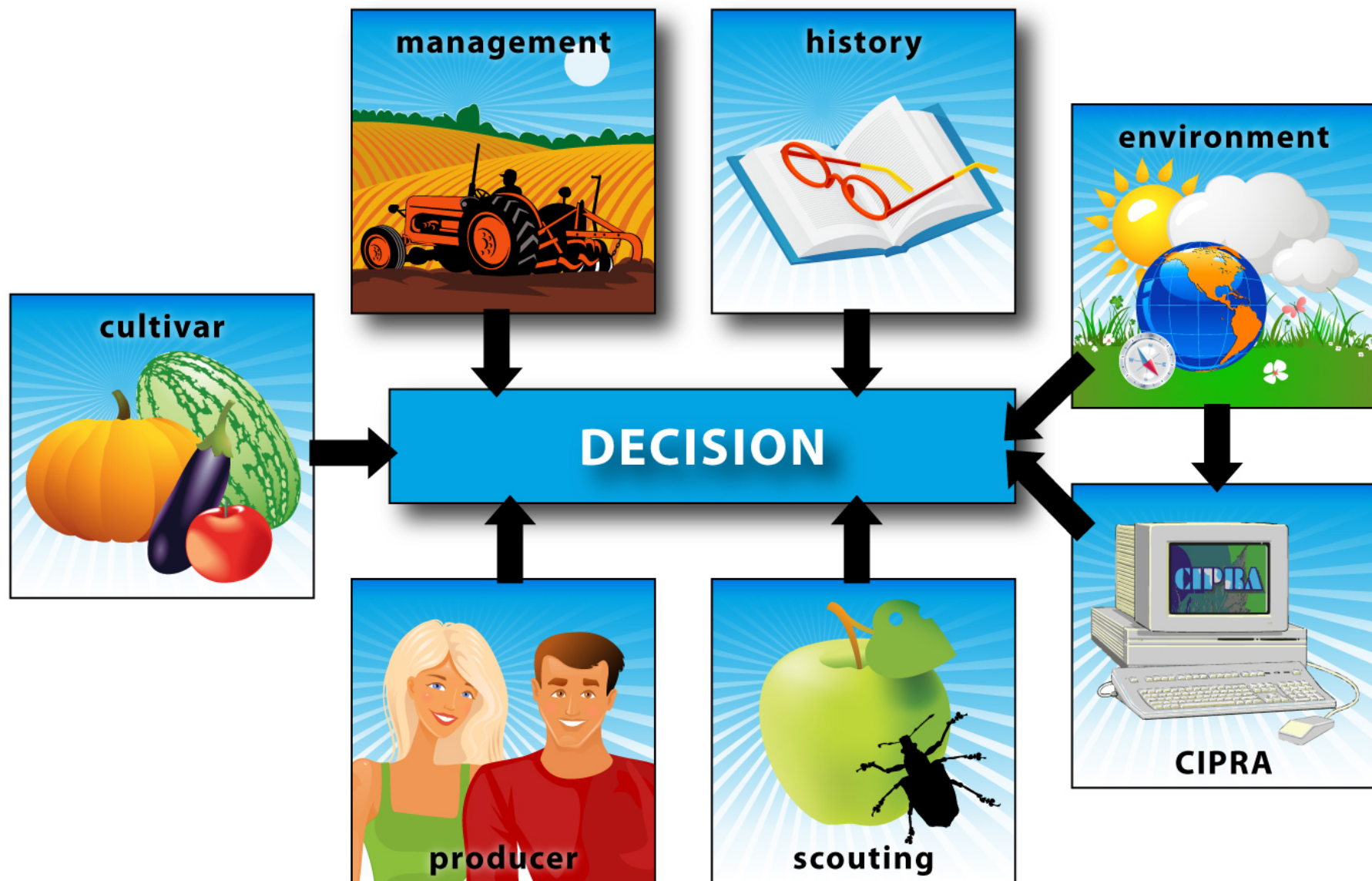


Bioclimatic modelling of crop-pest interactions to study the impacts of climate change and variability in Eastern Canada

Gaétan Bourgeois, Ph.D.

Saint-Jean-sur-Richelieu Research and Development Centre, QC

Canada 



Tair, Tsoil, RH, Wind,
SolRad, Precip

Tair, RH, Precip (%prob.),
Cloud Cover, Wind

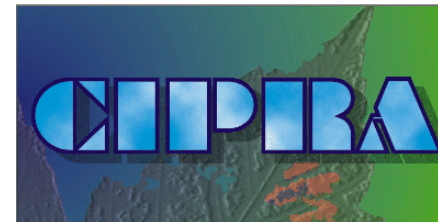
Tmax, Tmin, Precip

**Weather
observations
(hourly)**

**Weather
forecasts
(hourly)**

**Climate
normals
(daily)**

**Bioclimatic
models**



**Predicted
values**

**Biological
observations**

- Interpretation
- Evaluation
- Development and/or Update
- Technology transfer

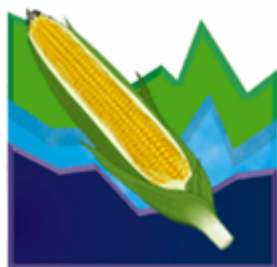
Computer Centre for Agricultural Pest Forecasting



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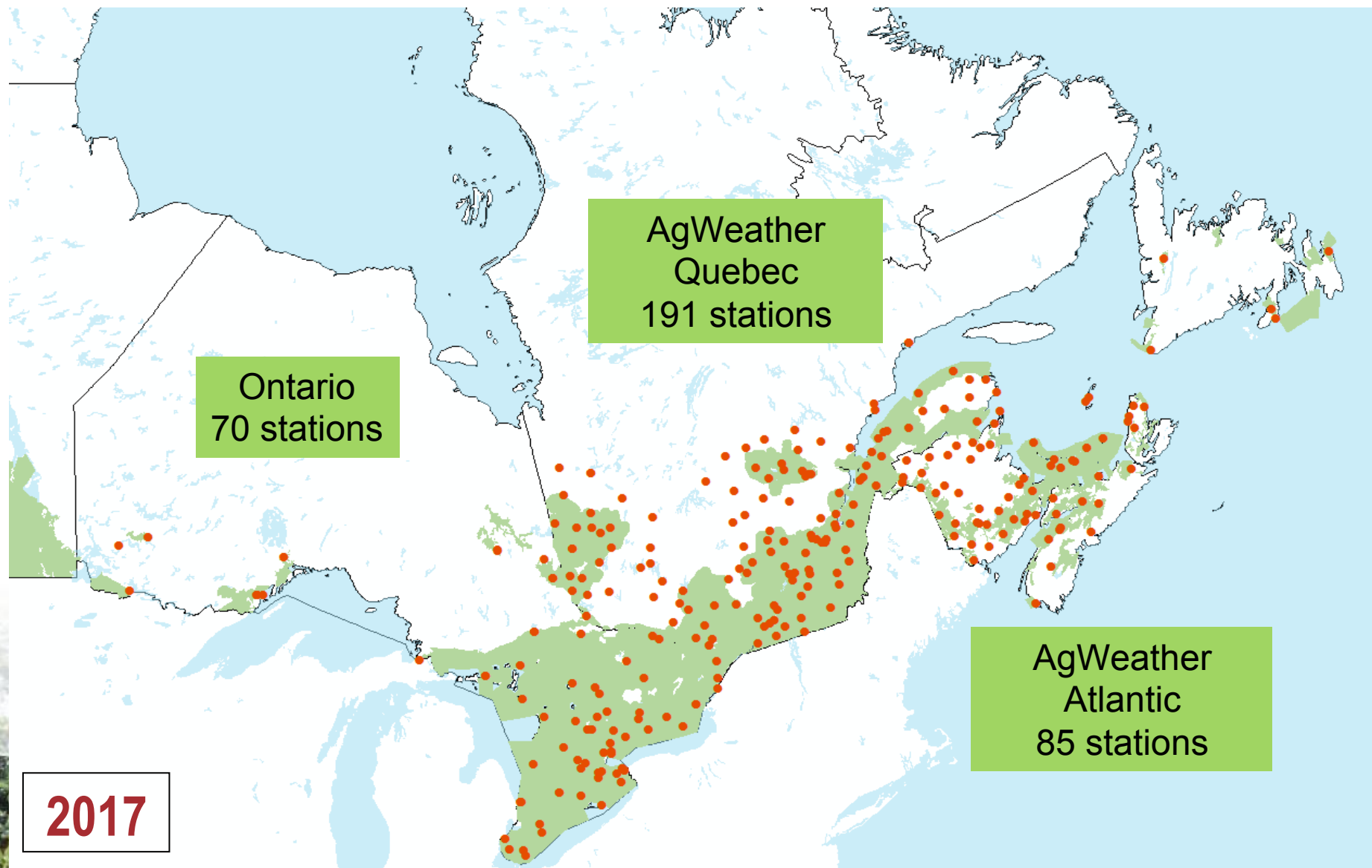


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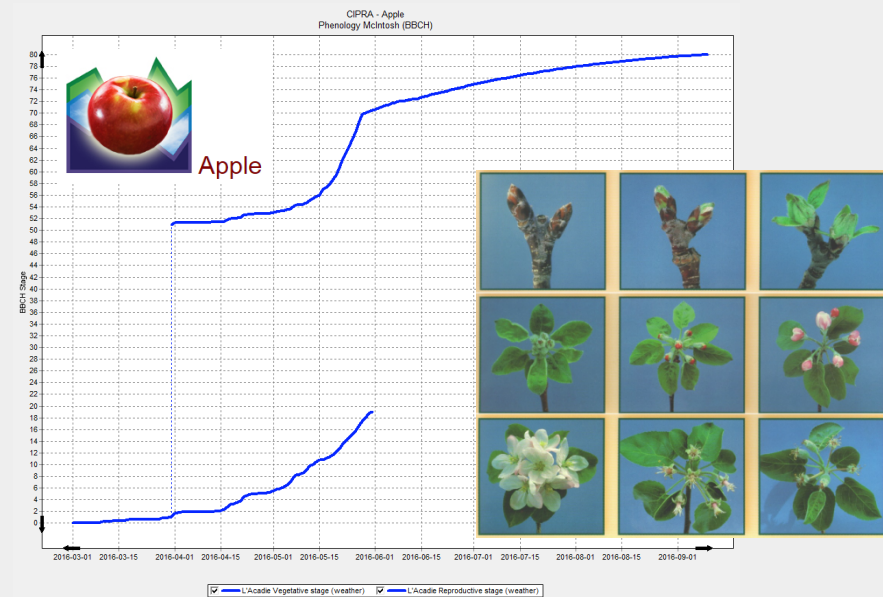
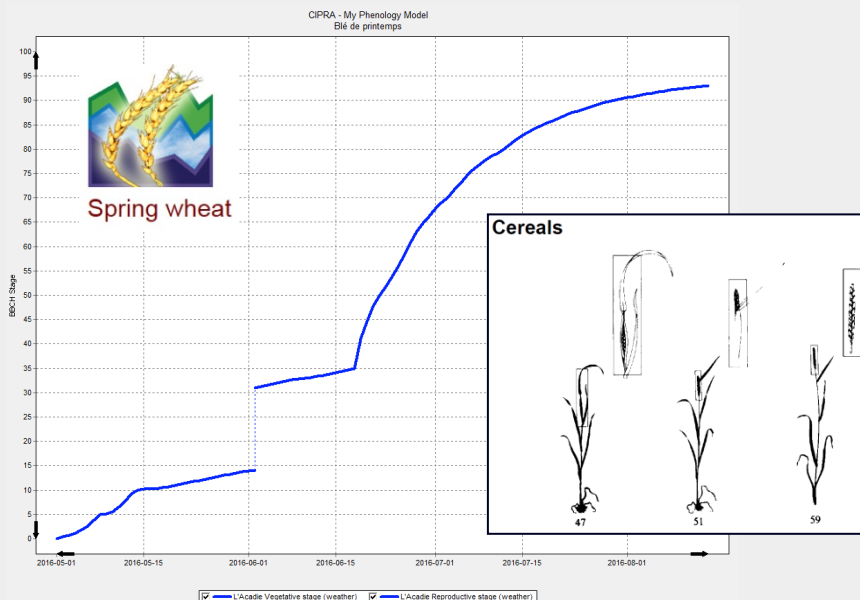
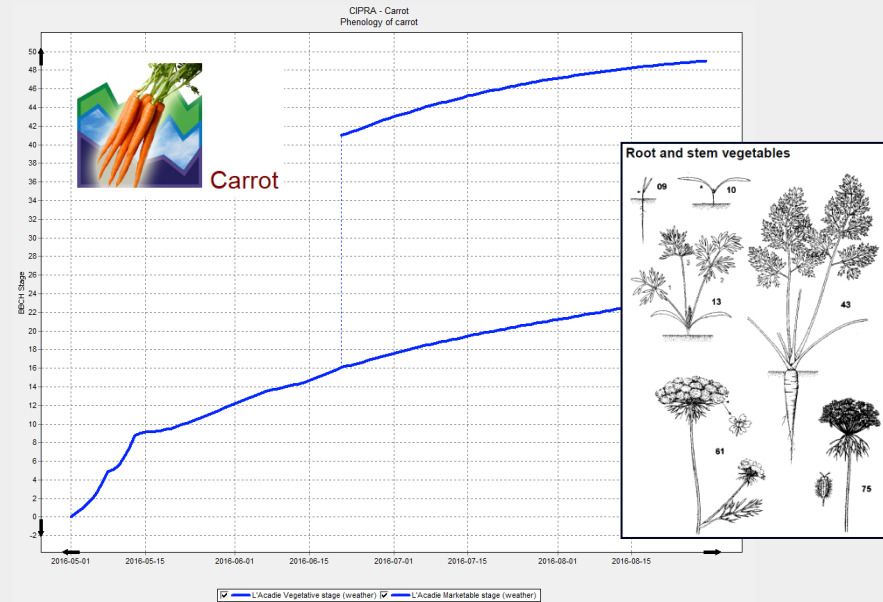
>130 Bioclimatic models and indices → >25 Crops
>250 Registered users → Thousands of crop producers

Operational weather network for bioclimatic models implemented in CIPRA for Eastern Canada (“real time”)

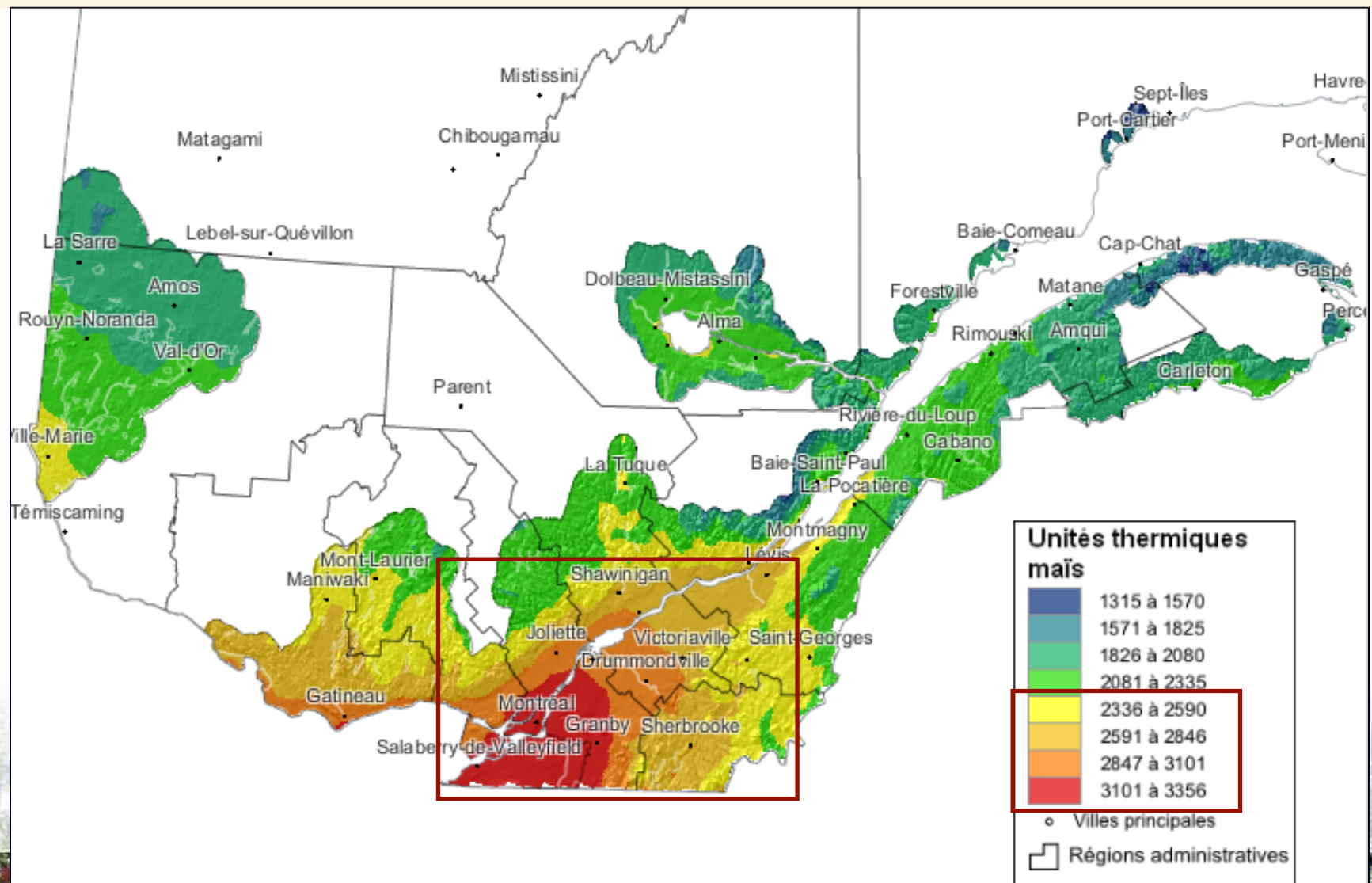


Predicting crop phenology (i.e. from seeding to harvest)

- Improves pest control and scouting strategies
- Better planning of seeding and crop harvest
- Great tool for marketing strategies
- Improves water and fertilizer management
- Key component for studies on impacts of climate change and variability



Cumulative corn heat units (CHU): 1979-2008



Préparé par :



En collaboration avec :



Agriculture et
Agroalimentaire Canada

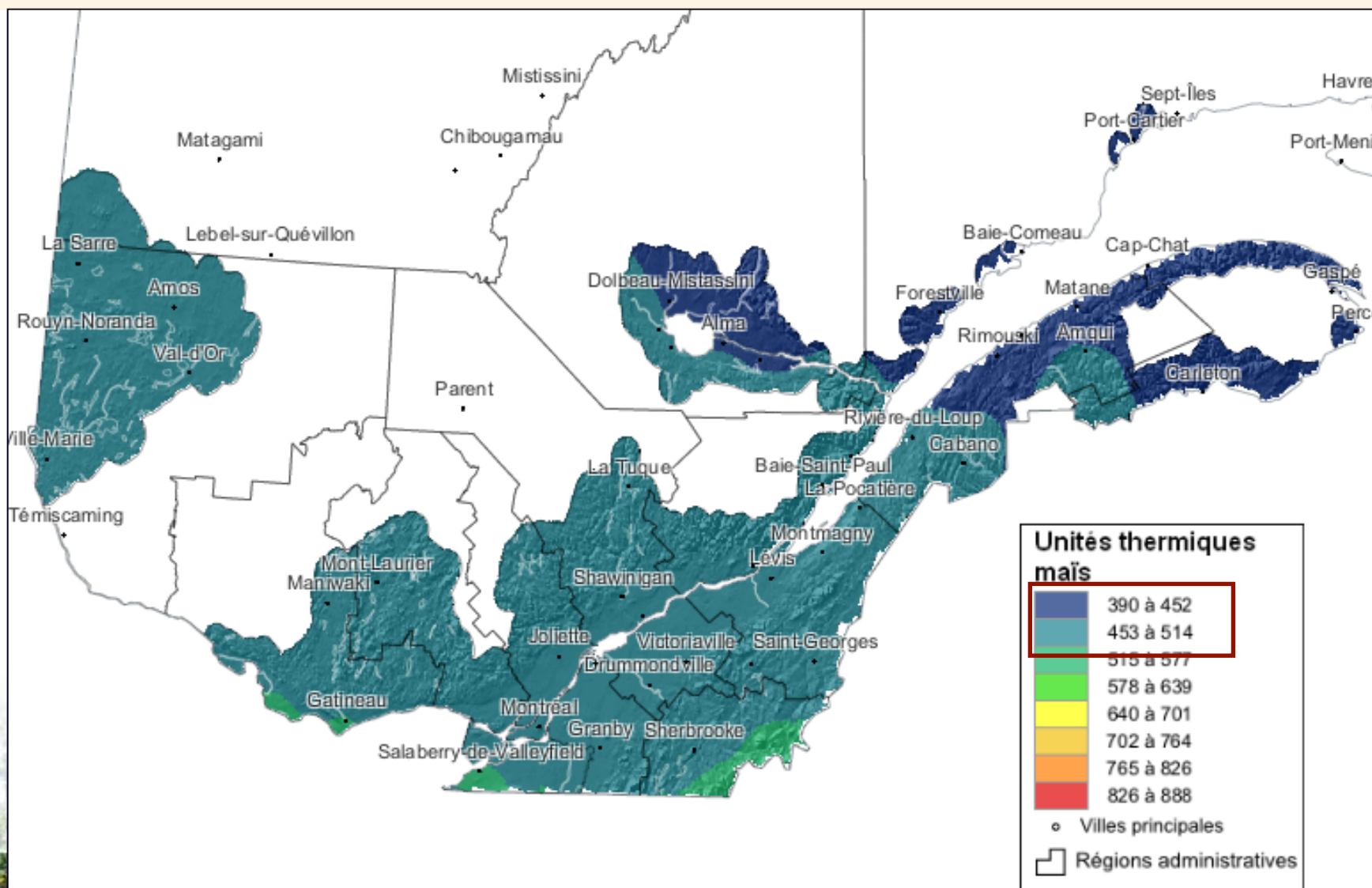
Agriculture and
Agri-Food Canada



Recherche scientifique
Canada

Canada

Change in corn heat units (CHU): 2041-2070 (Lower CC)



Préparé par :



En collaboration avec :



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Agroalimentaire Canada

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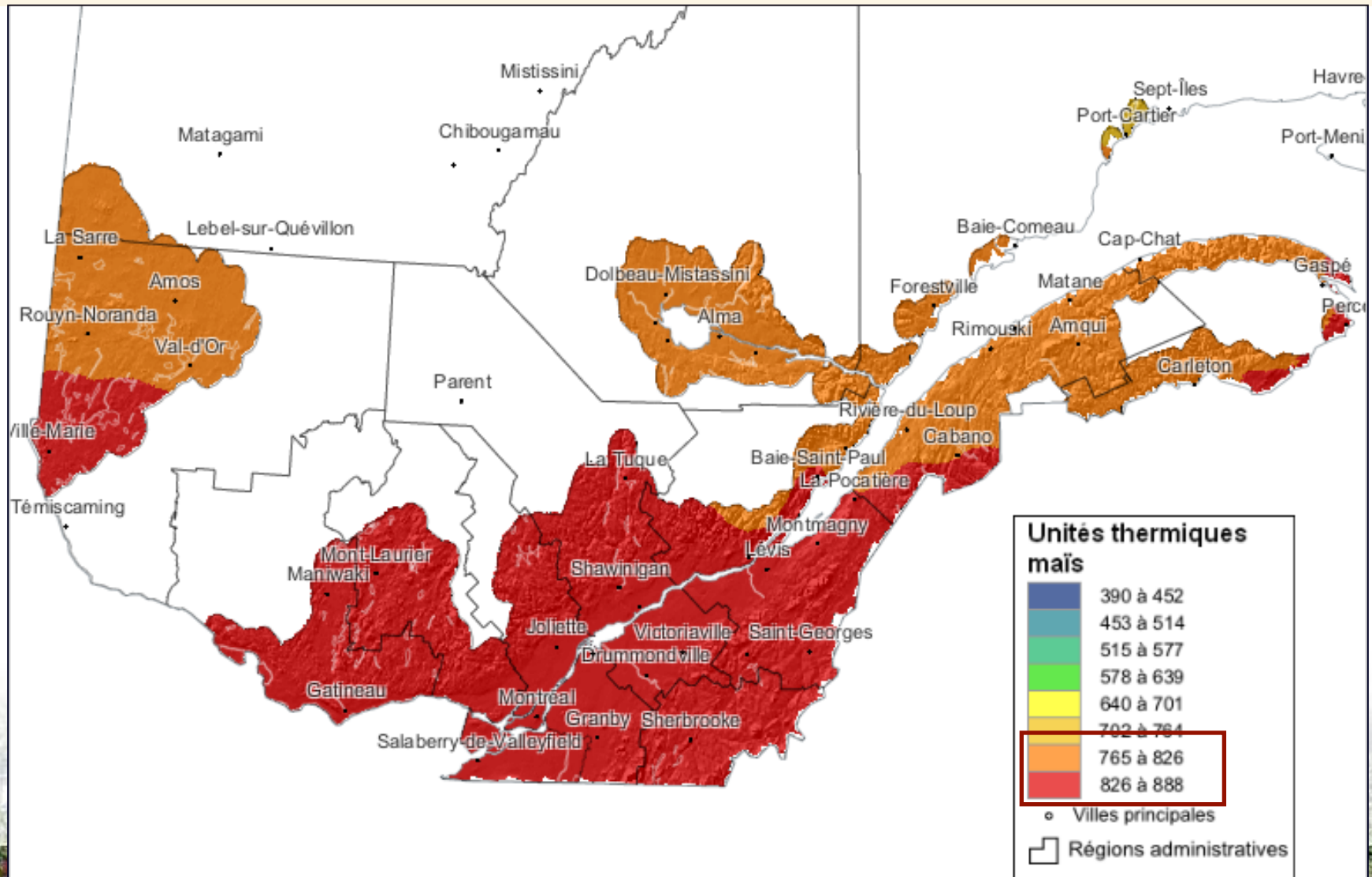
Québec



Ministère des Ressources naturelles
Canada

Canada

Change in corn heat units (CHU): 2041-2070 (Higher CC)



Préparé par :



En collaboration avec :



Agriculture et
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Québec



Recherches naturelles
Canada

Canada

WEATHER

M
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T

Crop canopy
structure

CROP

Phenology

Growth

Yield

- Productivity
- Quality

PESTS

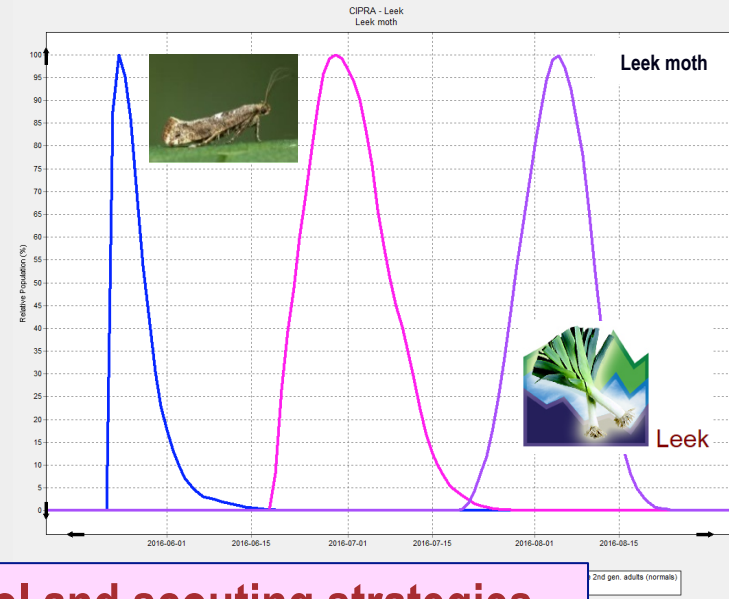
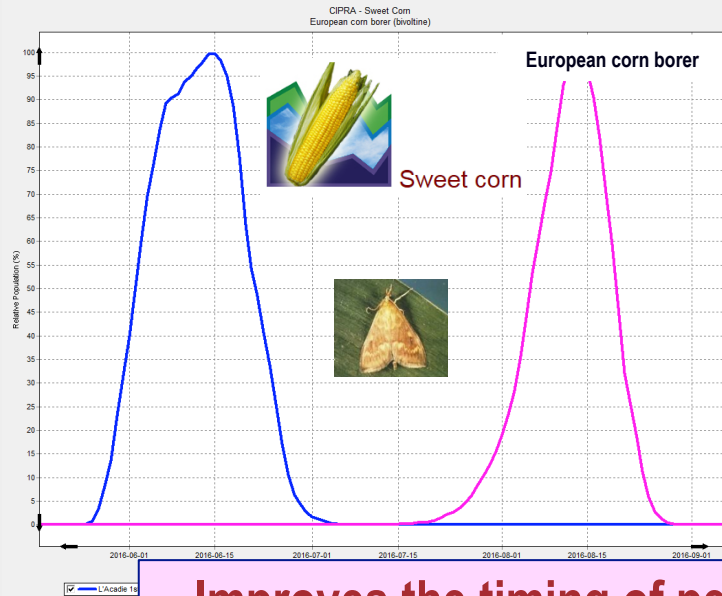
- Insects
- Diseases
- Nematodes
- Weeds



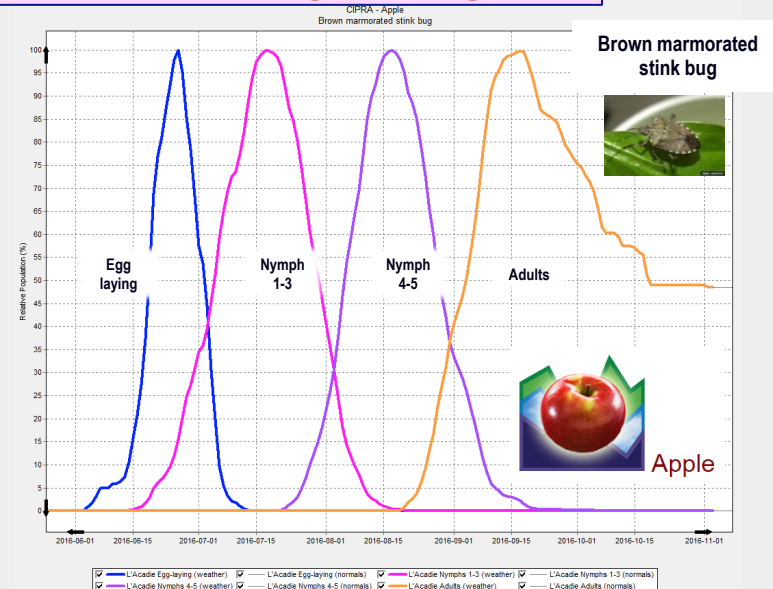
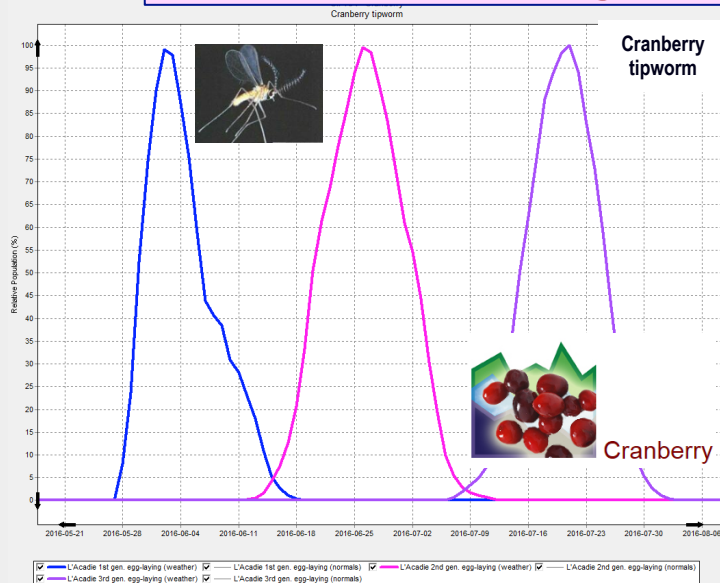
Pest forecasting:

- Improves timing of pest control and scouting strategies
- Decreases number of pesticide applications
 - Lower production costs
 - Positive impacts on the environment
 - Delays resistance development to pesticides

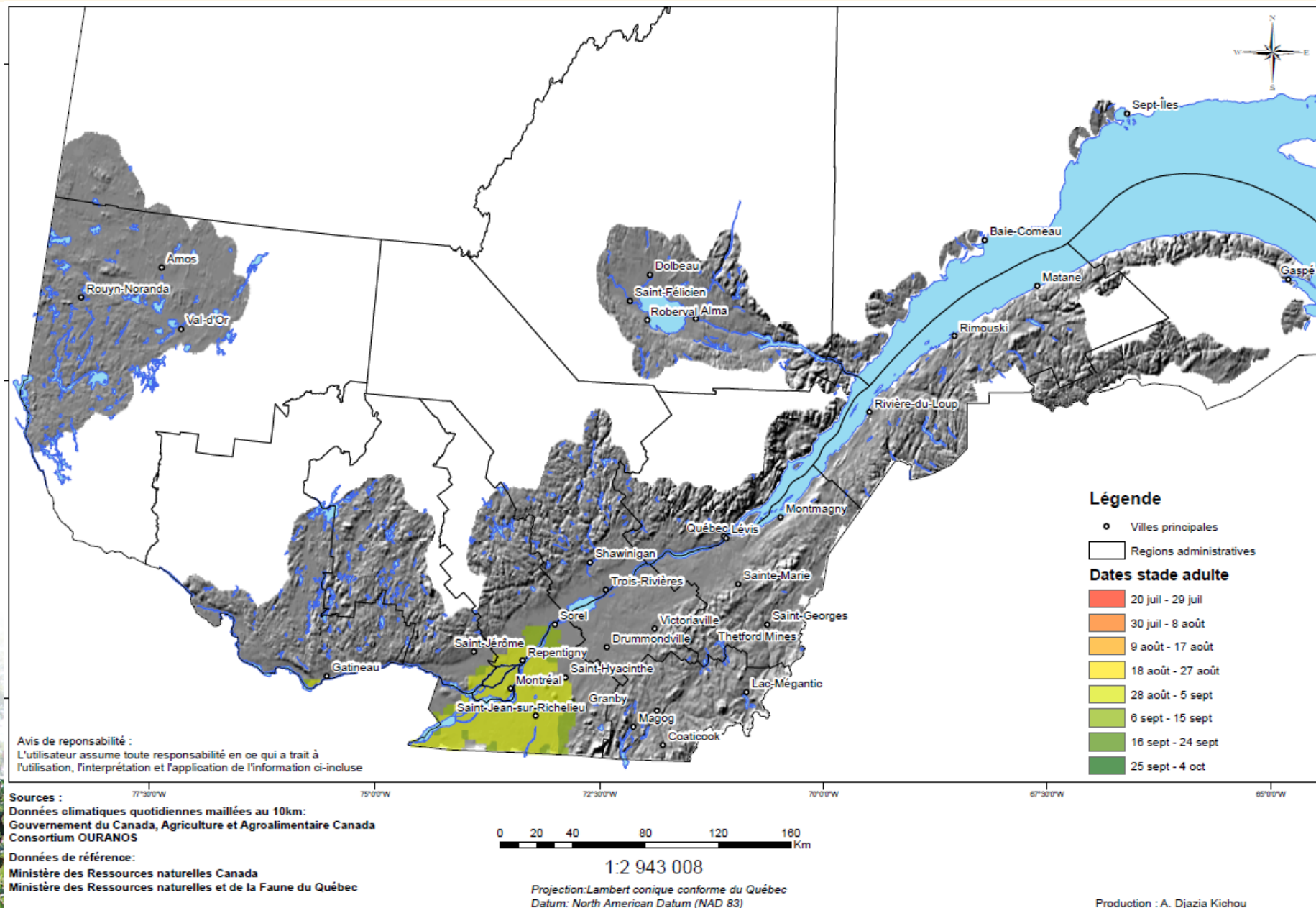
Predicting insect development for many crops



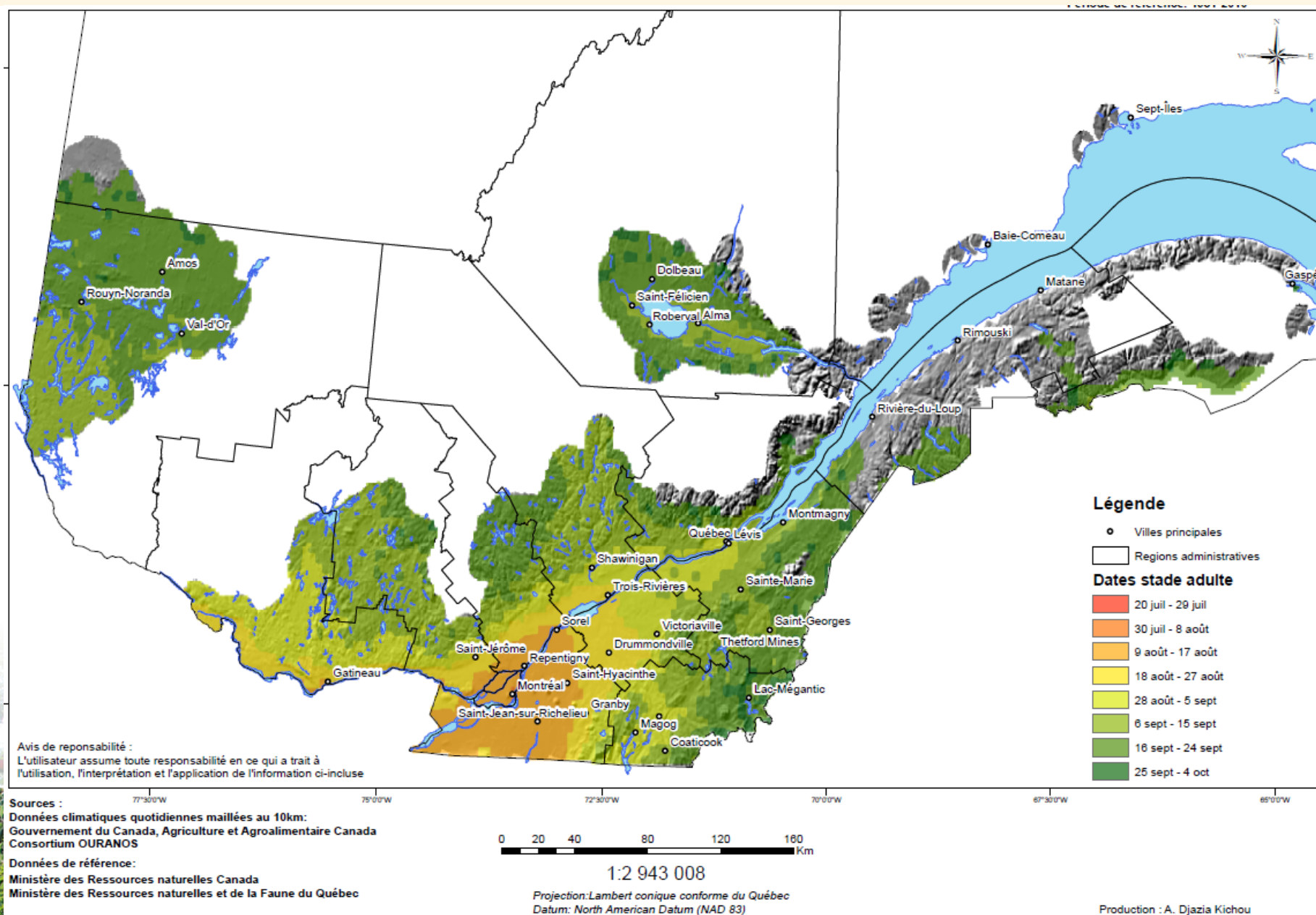
Improves the timing of pest control and scouting strategies



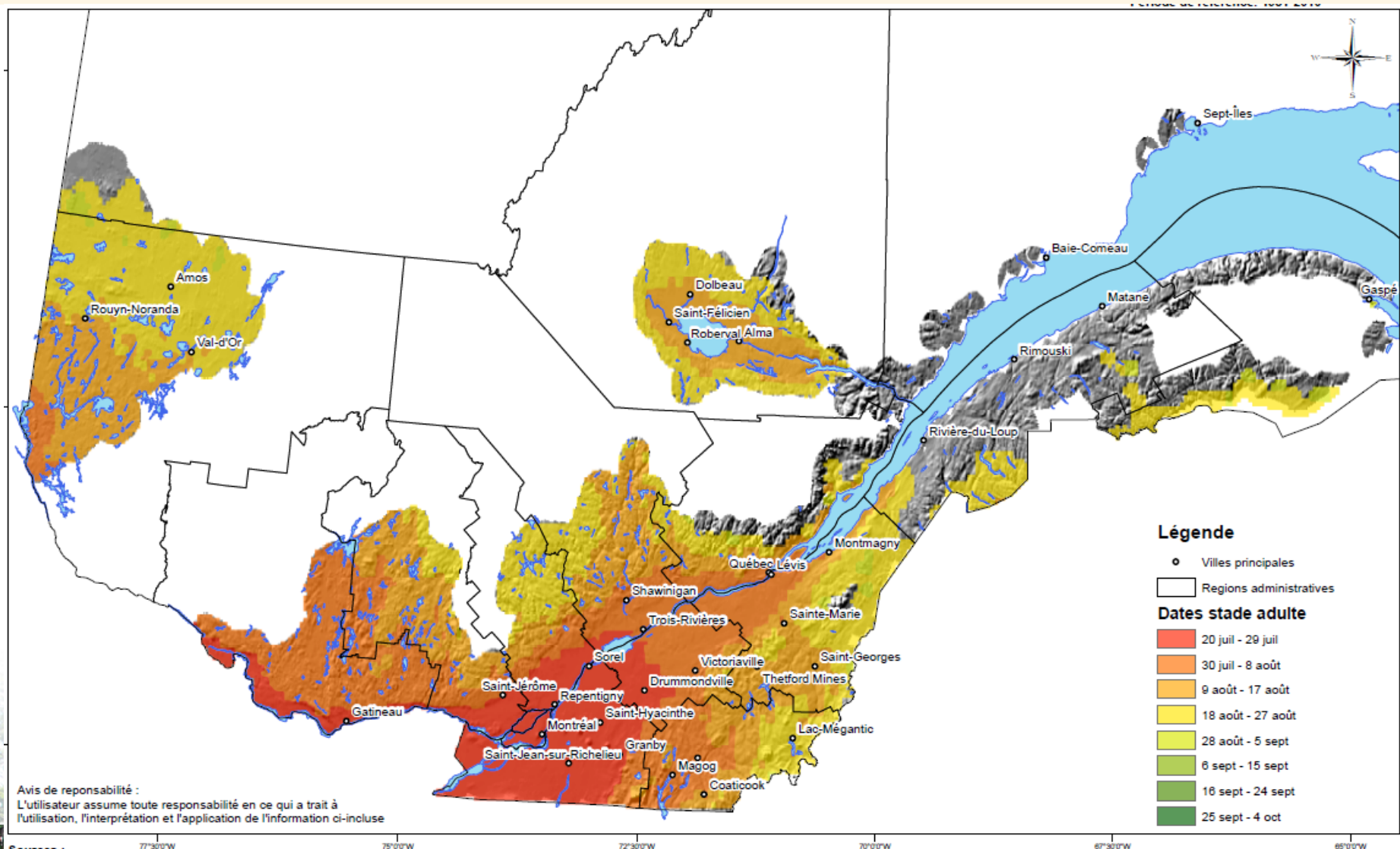
BMSB potential distribution in Quebec (1981-2010)



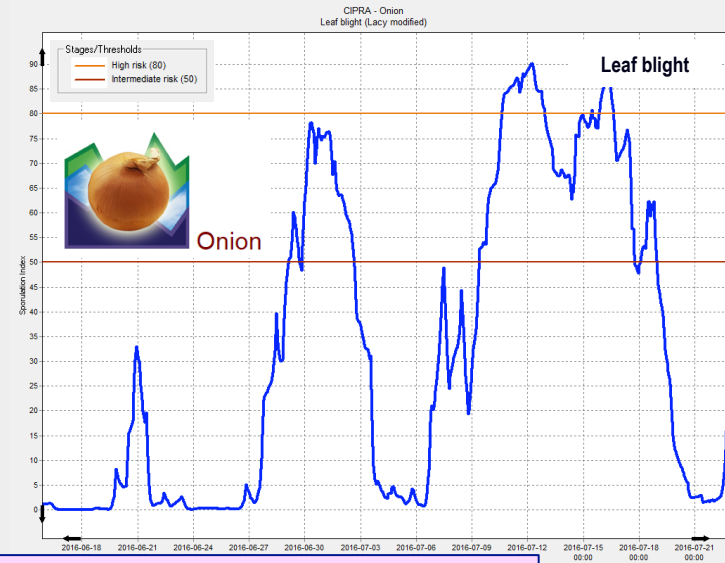
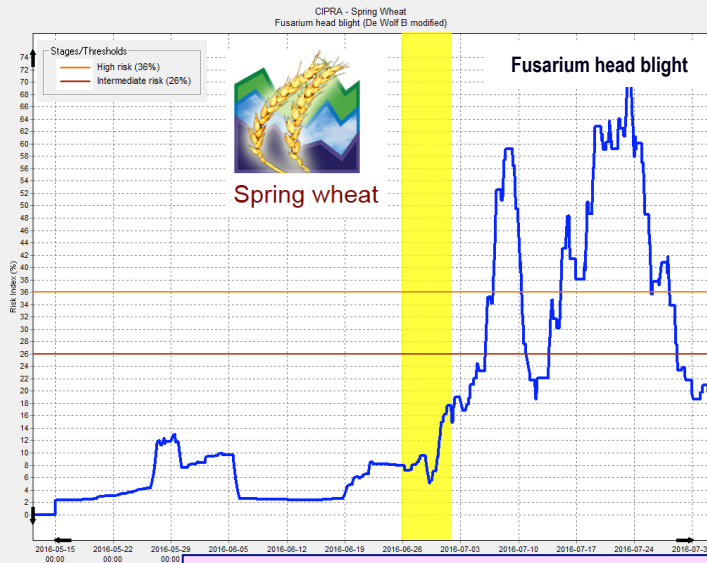
BMSB (Lower CC scenarios for 2041-2070)



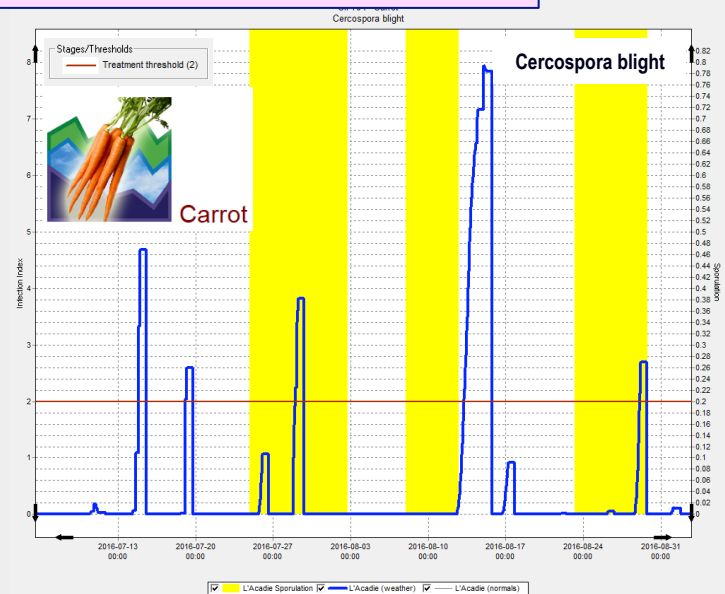
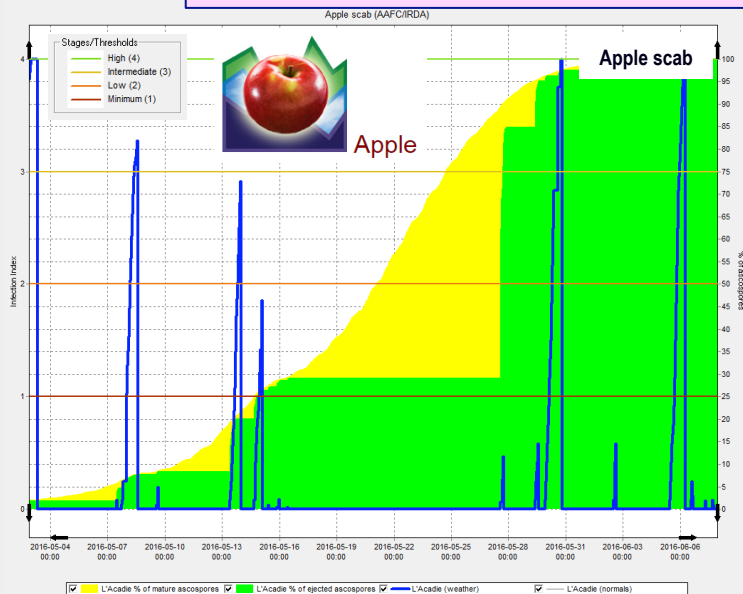
BMSB (Higher CC scenarios for 2041-2070)



Disease forecasting for many crops



Decreases the number of pesticide applications



Disease forecasting for many crops

Agriculture, Ecosystems and Environment 197 (2014) 147–158



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Climatic indicators for crop infection risk: Application to climate change impacts on five major foliar fungal diseases in Northern France



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Iñaki Garcia de Cortazar-Atauri^a, Marie-Odile Bancal^d, Nadine Brisson^{e,1}

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2.1.1. General surface response function development

The Weibull equation, as modified by Duthie (1997), was used to model the daily infection efficiency (hereafter “ClimInfer”) as a surface response to both temperature and LWD (Fig. 1, Eq. (1)):

$$\text{ClimInfer}(d) = f(T)(1 - \exp\{-[A(\text{LWD} - \text{LWD}_0)]^B\}) \quad (1)$$

$$\text{ClimInfer} = 0 \text{ for } T < 0$$

The simple temperature response function $f(T)$ is a linear-plateau function with ascending and descending portions on both sides of the optimal plateau (Hartkamp et al., 2002) (Eq. (2)):

$$f(T) = \left\{ \begin{array}{ll} 0, & T \leq T_{\min} \text{ and } T \geq T_{\max} \\ Y_{\max}, & T_{\text{opt}1} \leq T \leq T_{\text{opt}2} \\ \frac{Y_{\max}}{T_{\text{opt}1} - T_{\min}}(T - T_{\min}), & T_{\min} < T < T_{\text{opt}1} \\ \frac{Y_{\max}}{T_{\text{opt}2} - T_{\max}}(T - T_{\max}), & T_{\text{opt}2} < T < T_{\max} \end{array} \right\} \quad (2)$$

Disease forecasting for many crops

Table 1

Infection model parameters and evaluation by comparing model predictions with observations from published data.

Pathogen	Host	Disease	Ref. ^a	Estimated parameters [range from the literature]					Model evaluation			
				Y_{\max}^b	T_{\min}^c	$T_{\text{opt}1}^d$	$T_{\text{opt}2}^e$	T_{\max}^f	A^g	B^h	EF^i	$RRMSE^j$
<i>Leptosphaeria maculans</i>	Oilseed rape	Phoma leaf spot	(Toscano-Underwood et al., 2001)	0.311	3.33 [0;5]	15.15 [15;20]	16.28 [15;20]	29.83 [>20]	0.024	2.74	0.88	0.45
<i>Phytophthora infestans</i>	Potato	Late blight	(Maziero et al., 2009)	0.954	3.00 [0;11]	9.97 [10;24]	15.18 [10;24]	25.80 [25;28]	0.051	2.68	0.87	0.47
<i>Plasmopara viticola</i>	Grape	Downy mildew	(Lalancette et al., 1988)	0.085	4.79 [<5]	14.03 [15;24]	17.40 [15;24]	30.66 [30]	0.342	17.40	0.94	0.27
<i>Puccinia triticina</i>	Wheat	Leaf rust	(de Vallavieille-Pope et al., 1995)	0.436	1.86 [<5]	14.54 [12;20]	18.36 [12;20]	31.00 [28;30]	0.117	3.12	0.91	0.40
<i>Pyrenophora teres</i>	Barley	Net blotch	(Shaw, 1986)	0.345	1.94 [<4]	17.47 [13;18]	18.32 [13;18]	25.00 [25]	0.119	3.04	0.78	0.40

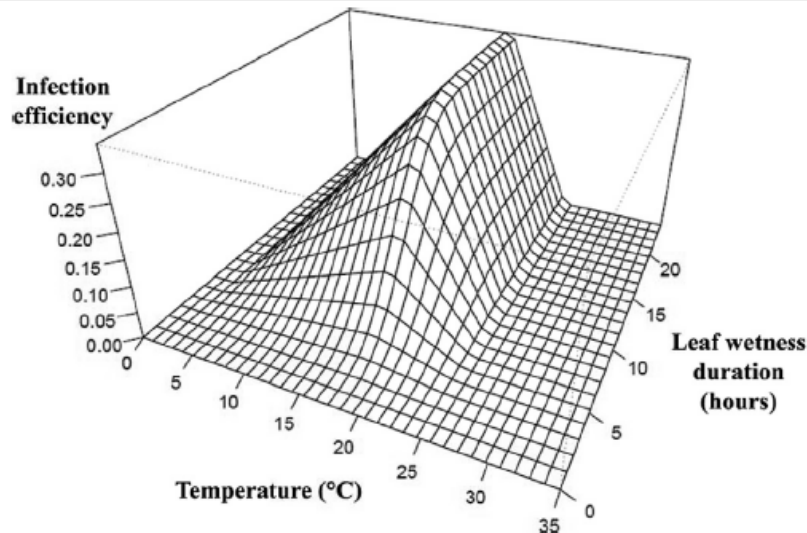


Fig. 1. Effects of temperature and leaf wetness duration on infection efficiency as predicted by the generic surface response function. This example shows the infection efficiency of *Puccinia triticina* ($Y_{\max}=0.436$, $A=0.117$, $B=3.12$, $T_{\min}=2^{\circ}\text{C}$, $T_{\text{opt}1}=15^{\circ}\text{C}$, $T_{\text{opt}2}=18^{\circ}\text{C}$, and $T_{\max}=31^{\circ}\text{C}$).

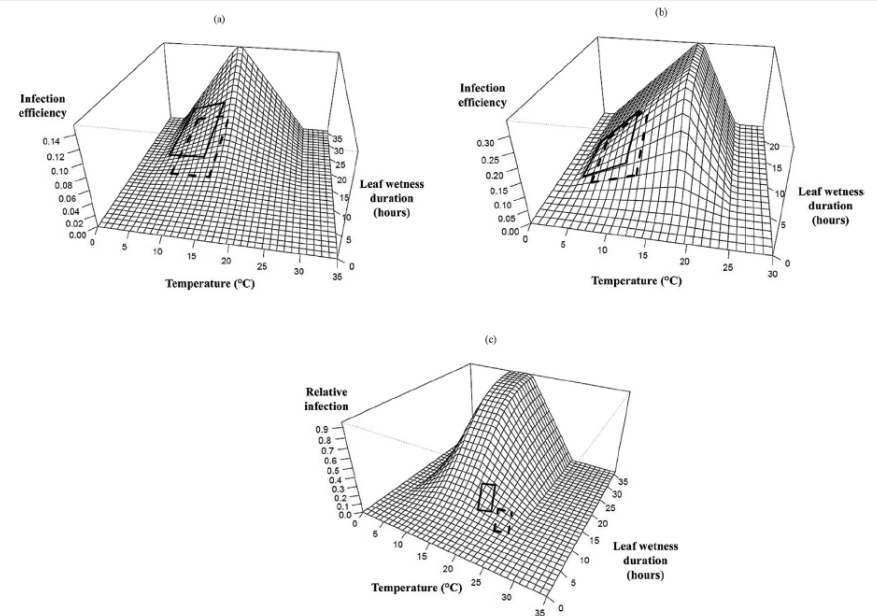
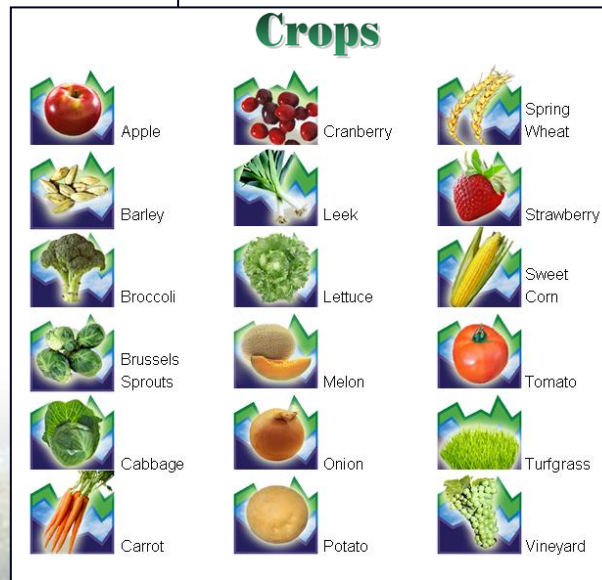


Fig. 3. Surface response of infection efficiency to temperature and leaf wetness duration for (a) phoma of oilseed rape, (b) net blotch of barley and (c) potato late blight. The rectangles depict the portions of the surface response corresponding to temperature and leaf wetness duration conditions in the recent past (—) and in the distant future (---) during the months of (a) October and November, (b) March–May and (c) June–August.

For more information on bioclimatic models within CIPRA

- Technical bulletin (Crop Guide) published in November 2014
- Available on-line
- Information
 - ✓ Descriptions
 - ✓ References
 - ✓ Interpretations
 - ✓ Calibration / Validation



Crop Guide



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CIPRA 2017:

A key tool for both

“Biovigilance” and “Precision horticulture”

Bioclimatology & Modelling Research Team

Saint-Jean-sur-Richelieu R&D Centre (QC)

For more information: Gaetan.Bourgeois@agr.gc.ca